

# Nonlinear and Stochastic Morphological Segregation

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**Abstract.** I perform a joint counts-in-cells analysis of galaxies of different spectral types using the Las Campanas Redshift Survey (LCRS). Using a maximum-likelihood technique to fit for the relationship between the density fields of early- and late-type galaxies, I find a relative linear bias of  $b = 0.76 \pm 0.02$ . This technique can probe the nonlinearity and stochasticity of the relationship as well. However, the degree to which nonlinear and stochastic fits improve upon the linear fit turns out to depend on the redshift range in question. In particular, there seems to be a systematic difference between the high- and low-redshift halves of the data (respectively, further than and closer than  $cz \approx 36,000$  km/s); all of the signal of stochasticity and nonlinearity comes from the low-redshift portion. Analysis of mock catalogs shows that the peculiar geometry and variable flux limits of the LCRS do not cause this effect. I speculate that the central surface brightness selection criteria of the LCRS may be responsible.

## 1. Introduction

Different types of galaxies have different large-scale density fields (Hubble 1936; Oemler 1974). Various authors have compared elliptical and spiral galaxies, finding that the fluctuation amplitude of ellipticals is stronger than that of spirals by a factor of 1.3–1.5 (Santiago & Strauss 1992; Loveday *et al.* 1996; Hermit *et al.* 1996; Guzzo *et al.* 1997). Similarly, a comparison of the galaxy distribution in the *IRAS* redshift survey (Strauss *et al.* 1992b) with those in the Center for Astrophysics Redshift Survey (CfA; Huchra *et al.* 1983) and in the Optical Redshift Survey (ORS; Santiago *et al.* 1995), shows that optically-selected galaxies are clustered more strongly than infrared-selected galaxies by a similar factor (Davis *et al.* 1988; Babul & Postman 1990; Strauss *et al.* 1992a).

All of these statistics, however, only compare the *amplitudes* of the fluctuations of each galaxy type. It may be that the pattern of fluctuations is different for different galaxy types, as well. That is, that the local density of early-type galaxies is not deterministically related to the local density of late-type galaxies; other variables may affect the relationship. One way to quantify this is by using the correlation coefficient  $r_{el} \equiv \langle \delta_e \delta_l \rangle / \sigma_e \sigma_l$ , where  $\delta$  refers to the galaxy overdensity and  $\sigma^2 \equiv \langle \delta^2 \rangle$  is its variance.  $r_{el} = 1$  if the fields are perfectly correlated, and  $r_{el} = 0$  if they are completely uncorrelated. In large-volume hydrodynamic simulations of galaxy formation, Blanton *et al.* (1999) have found a low corre-

lation coefficient between the density fields of old and young galaxies, around  $r_{el} \sim 0.5$ . The differences arise in large part from the fact that the dependence of each galaxy density field on temperature is different; young galaxies are not found in high-temperature regions in the simulations.

Motivated by these theoretical results, in this proceeding I perform a point-by-point comparison of the density fields of different galaxy types, using a counts-in-cells analysis of the Las Campanas Redshift Survey (LCRS; Shectman *et al.* 1996). The maximum-likelihood method I use is superior to the comparison of second moments alone (Tegmark & Bromley 1999), as I will demonstrate below, and can probe the nonlinearity and stochasticity of the relationship between different galaxy types. I describe my method in Section 2, describe the nature of the data in Section 3, describe my results in Section 4, and show results from mock catalogs in Section 5. The results presented at the meeting have been amended; in particular, the stochasticity detected between the density fields turns out to depend critically on redshift, as explained more completely below. I conclude in Section 6.

## 2. Method

I am interested in  $f(\delta_e, \delta_l)$ , the joint distribution of the overdensities  $\delta_e$  of early-type galaxies and  $\delta_l$  of late-type galaxies. Dekel & Lahav (1999) show how such a joint distribution determines the relationship between the correlation functions, the power spectra, and higher-order moments of two density fields. It is possible to constrain the properties of  $f(\delta_e, \delta_l)$  using these quantities. However, a more direct approach is to analyze the related joint probability distribution  $P(N_e, N_l)$  of finding  $N_e$  early-type and  $N_l$  late-type galaxies in a single cell of size  $R$ . After all, this latter probability is simply  $f(\delta_e, \delta_l)$  (where the fields are smoothed over the volume of a cell) convolved with Poisson distributions. If one notes that  $f(\delta_e, \delta_l) = f(\delta_l|\delta_e)f(\delta_e)$ , one can write

$$P(N_e, N_l|\alpha) = \int d\delta_e \frac{\bar{N}_e^{N_e} (1 + \delta_e)^{N_e}}{N_e!} e^{-\bar{N}_e(1+\delta_e)} f(\delta_e) \times \int d\delta_l \frac{\bar{N}_l^{N_l} (1 + \delta_l)^{N_l}}{N_l!} e^{-\bar{N}_l(1+\delta_l)} f(\delta_l|\delta_e), \quad (1)$$

where  $\bar{N}_e$  and  $\bar{N}_l$  are the average number of galaxies of each type expected in a cell of a given volume (and given selection criteria), and  $\alpha$  represents the parameters of  $f(\delta_e, \delta_l)$ . One can then fit for  $\alpha$  by minimizing the quantity  $\mathcal{L} \equiv -2 \sum_i \ln P(N_{e,i}, N_{l,i}|\alpha)$ . In practice, I first fit for the parameters of the density distribution  $f(\delta_e)$ ; then I fix the parameters of  $f(\delta_e)$  and fit separately for those of  $f(\delta_l|\delta_e)$ . I calculate error bars using the well-known bootstrap method.

I parameterize  $f(\delta_l, \delta_e)$  by first specifying the density distribution function  $f(\delta_e)$  and then the bias relation  $f(\delta_l|\delta_e)$ . Here, I will use a log-normal model for the density distribution function:

$$f(\delta_e) = \frac{1}{\sqrt{2\pi}\sigma_e(1 + \delta_e)} \exp \left[ -x_e^2 / 2\sigma_e^2 \right]. \quad (2)$$

where  $x_e \equiv \ln(1 + \delta_e) + \sigma_e^2/2$ . This model is a better fit to the data than either a Gaussian or a first-order Edgeworth expansion (Bernardeau & Kofman 1994; Juszkiewicz *et al.* 1995; Kim & Strauss 1998); however, the results presented here do not depend sensitively on my choice for  $f(\delta_e)$ .

The bias relation  $f(\delta_l|\delta_e)$  can be either deterministic or stochastic. These terms are not meant to refer to underlying physical principles but are simply meant to express whether knowing the density of ellipticals tells you with certainty the density of spirals, modulo Poisson noise. In the case of deterministic bias, the joint density distribution can be expressed as

$$f(\delta_l|\delta_e) = \delta^D(\delta_l - b(\delta_e)) \quad (3)$$

where  $f(\delta_e)$  is the density distribution function of early-type galaxies. Under this assumption, one can describe the models by the function  $b(\delta_e)$ .

The simplest model for  $b(\delta_e)$  is linear bias:

$$b(\delta_e) = b_0 + b_1\delta_e, \quad (4)$$

which can be trivially extended to quadratic bias:

$$b(\delta_e) = b_0 + b_1\delta_e + b_2\delta_e^2. \quad (5)$$

Another possibility is “broken” bias, which is piece-wise linear with one slope in overdense regions and another in underdense regions:

$$b(\delta_e) = \begin{cases} b_0 + b_1\delta_e & \text{for } \delta_e < 0 \\ b_0 + b_2\delta_e & \text{for } \delta_e > 0 \end{cases} \quad (6)$$

I require that  $\langle \delta_l \rangle = 0$ , because it is meant to represent the overdensity of late-type galaxies. In practice, this requirement sets  $b_0$ , which is therefore not treated as a free parameter in any of the above expressions.

If variables other than the local density field are important in determining where galaxies form, it may be that the different formation processes of early-type and late-type galaxies cause scatter in their relationship. Thus, I examine models which incorporate scatter by rewriting Equation (3), replacing the Dirac delta-function with a Gaussian of finite width:

$$f(\delta_l|\delta_e) = \frac{1}{\sqrt{2\pi}\sigma_b} \exp \left[ -\frac{(\delta_l - b(\delta_e))^2}{2\sigma_b^2} \right] \quad (7)$$

where  $b(\delta_e)$  and  $f(\delta_e)$  are chosen as above. In the case that  $f(\delta_e)$  is Gaussian, and  $\sigma$  and  $\sigma_b$  are small, such that the limit  $\delta \geq -1$  is not important,  $f(\delta_e, \delta_l)$  reduces to a bivariate Gaussian distribution, and the standard correlation coefficient is related to  $\sigma_b$  by

$$r = \sqrt{1 - (\sigma_b/\sigma_l)^2} \quad (8)$$

I will use Equation (7) to fit linear bias with Gaussian scatter to the relationship between galaxy types.

### 3. Galaxies in the LCRS

The LCRS (Shectman *et al.* 1996) consists of  $\sim 25,000$  galaxies with a median redshift of  $z \sim 0.1$ . Three long slices ( $1.5^\circ \times 80^\circ$ ) were surveyed in the North Galactic Cap, and three in the South Galactic Cap. Within each hemisphere, the slices had the same right-ascension limits but were separated by several degrees in declination. In addition to the flux limits, a cut was applied on the central magnitude (the magnitude within the central two pixels of the CCD images), to avoid putting fibers onto galaxies unlikely to yield redshifts. As in all redshift surveys, this cut can affect the relationship between the luminosity function and the selection function, since the selection is not purely based on flux.

Bromley *et al.* (1998) have used a spectral classification scheme to divide the galaxies into six “clans.” For my purposes, I will split the galaxies into just two groups: an early-type group consisting of clans 1 and 2 and a late-type group consisting of clans 3 through 6. I place absolute magnitude limits on the early-type group of  $-22.5 < M < -18.8$  and on the late-type group of  $-22.0 < M < -18.5$ . This procedure yields about 10,000 galaxies in each group.

The geometry of the LCRS complicates an attempt to perform a counts-in-cells analysis on it. I create 14 redshift shells, each with an equal volume; thus, the shells at higher redshift have a shorter radial extent. In the angular dimension, I divide the survey into cells which are 3 spectrograph fields on each side (each field is about  $1.5^\circ \times 1.5^\circ$ ). In the right ascension direction, the fields are adjacent; in the declination direction, the fields from the three slices in each Galactic hemisphere are combined. This procedure produces 518 cells total, each with a volume equivalent to a  $15 h^{-1}$  Mpc sphere, of about cubical dimensions at  $z \sim 0.1$  (except for the gaps in the declination direction).

### 4. Results from the LCRS

I derive the luminosity and selection functions from the data and calculate the expected counts in each cell for each galaxy type. The distribution of  $N/N_{\text{exp}}$  is shown in Figure 1 for each galaxy type. The top quarter of Table 1 shows the results of fitting for the log-normal distribution and each bias model. Note first that for the linear fit  $b_1 = 0.76 \pm 0.02$ ; that is, the late-type galaxies are unbiased with respect to the early-type galaxies, to a degree which agrees with conventional wisdom. Some nonlinearity is detected in the quadratic bias case, at rather high significance ( $6\sigma$ ), in the sense that the slope of the bias steepens at large densities (*i.e.*  $b_2 > 0$ ). The broken bias model also shows this effect, at somewhat lower significance. Stochasticity is detected at about  $10\sigma$  significance, with  $\sigma_b = 0.21 \pm 0.02$ . This level of stochasticity corresponds to  $r \approx 0.87$ ; for comparison, the moments method of Tegmark & Bromley (1999) would estimate  $r$  for this counts-in-cells distribution to be  $r = 0.73 \pm 0.01$ . Apparently a considerable amount of the “stochasticity” measured by the moments method is due to an inadequate characterization of the distribution of the densities, most likely because the method does not account for the non-Gaussianity of the Poisson distribution at low  $N$  or for the lower limit of  $\delta \geq -1$ .

The likelihoods relative to the linear fit are also shown in Table 1 for all the other models, and indicate that the stochastic linear bias is clearly the best

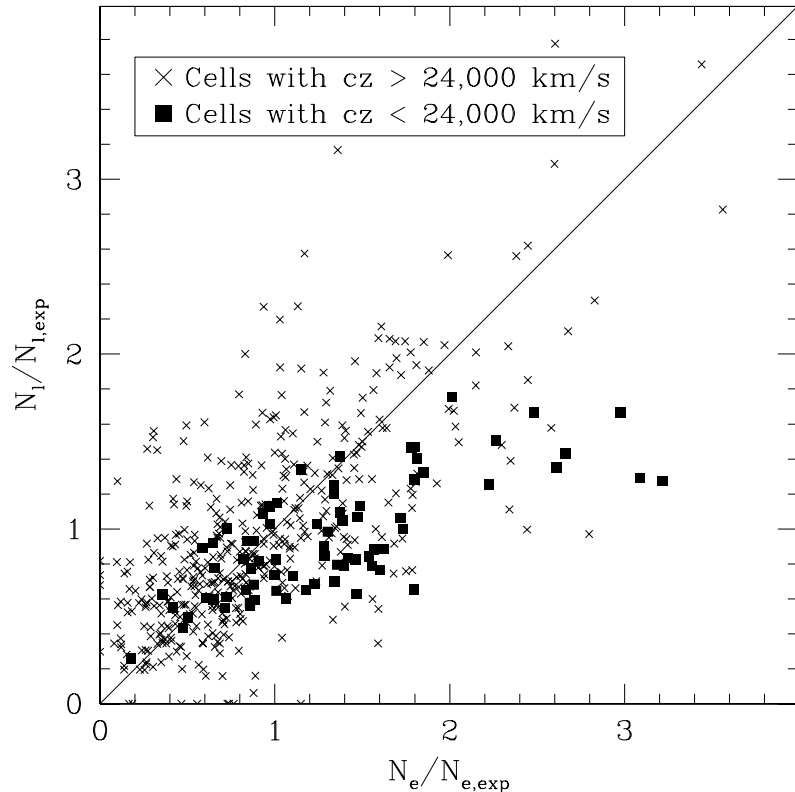


Figure 1. Joint overdensity distribution of early-type galaxy ( $x$ -axis) and late-type galaxy ( $y$ -axis) counts-in-cells for the LCRS. The low-redshift cells are shown as solid squares and the high-redshift cells are shown as crosses. Note that it appears as if the low-redshift cells are systematically less biased than the high redshift cells, which accounts for the increased stochasticity and nonlinearity when these cells are included. This low bias could occur because the selection function for the late-type galaxies is overestimated at low redshifts.

model I have considered. I have also compared each of the two-parameter fits to the linear fit by using a likelihood ratio test, whose results I show in the last column as the probability  $P_{\text{random}}$  of getting the likelihood difference between the linear model and the two-parameter model by chance. Note that since I only ran 200 realizations for each estimate, there is a lower limit on  $P_{\text{random}}$  of 0.005. From these results, it is clear that I detect nonlinearity at a statistically significant level, and stochasticity at an extremely significant level.

As part of an effort to test for systematic effects, I split the cells into a high-redshift half and a low redshift half. Results for each half are shown in the middle two quarters of Table 1. I find that the low redshift half retains the large amount of nonlinearity and stochasticity, while the high redshift half has no detectable signal. I can be more precise and eliminate the two innermost rings of cells, and fit to the rest, as shown in the bottom quarter of Table 1. This set of cells shows no nonlinearity, and a much reduced stochasticity, indicating that most of the signal for stochastic bias was coming from the two innermost

Table 1. Properties of PDF and bias fits to LCRS joint counts-in-cells of early and late type galaxies, for different redshift selections.

$cz$ Range (km/s)	$\sigma_e$	$\sigma_l$	Bias Model	$b_1$	$b_2$ or $\sigma_b$	$\mathcal{L}$	$P_{\text{random}}$
10,000–46,000	$0.54 \pm 0.02$	$0.41 \pm 0.02$	Linear	$0.76 \pm 0.02$	—	0.0	N/A
			Quadratic	$0.73 \pm 0.03$	$0.18 \pm 0.03$	−13.1	< 0.005
			Broken	$0.64 \pm 0.05$	$0.89 \pm 0.05$	−5.2	0.020
			Stochastic	$0.63 \pm 0.05$	$0.21 \pm 0.02$	−106.0	< 0.005
10,000–36,000	$0.50 \pm 0.03$	$0.39 \pm 0.02$	Linear	$0.75 \pm 0.03$	—	0.0	—
			Quadratic	$0.73 \pm 0.04$	$0.19 \pm 0.05$	−8.2	0.005
			Broken	$0.66 \pm 0.05$	$0.85 \pm 0.06$	−1.8	0.165
			Stochastic	$0.55 \pm 0.06$	$0.22 \pm 0.02$	−119.7	< 0.005
36,000–46,000	$0.61 \pm 0.03$	$0.46 \pm 0.03$	Linear	$0.79 \pm 0.05$	—	0.0	—
			Quadratic	$0.81 \pm 0.06$	$−0.06 \pm 0.05$	−0.6	0.420
			Broken	$0.80 \pm 0.11$	$0.78 \pm 0.08$	−0.0	0.955
			Stochastic	$0.78 \pm 0.04$	$0.07 \pm 0.05$	−0.2	0.370
24,000–46,000	$0.56 \pm 0.03$	$0.43 \pm 0.03$	Linear	$0.81 \pm 0.03$	—	0.0	—
			Quadratic	$0.81 \pm 0.04$	$0.00 \pm 0.03$	−0.0	0.990
			Broken	$0.82 \pm 0.07$	$0.80 \pm 0.05$	−0.0	0.920
			Stochastic	$0.77 \pm 0.04$	$0.13 \pm 0.03$	−8.9	< 0.005

rings of cells; tests show that this effect is *not* due to the better signal-to-noise in the inner rings. Again I can express the stochasticity in terms of the correlation coefficient  $r \approx 0.95$ ; the moments method of Tegmark & Bromley (1999) obtains  $r = 0.93 \pm 0.03$  for this set of cells.

In order to understand this effect, reconsider Figure 1, where I have marked the cells in the inner two rings with square boxes. It is clear that these cells have a smaller bias compared to the other cells. This probably indicates that the selection function is overestimated for late-type galaxies at low redshifts. I have performed the same analyses using selection functions based on the luminosity functions of Bromley *et al.* (1998), and find the same effect.

## 5. Results from Mock Catalogs

Because of the peculiar geometry and selection effects of the LCRS it is necessary to test these results against mock catalogs where I have simulated the observational effects inherent in the survey. I would also like to understand whether some of the systematic trends with redshift found in the last section can be explained by observational effects. I run particle-mesh simulations of a  $300 h^{-1}$  Mpc box using  $256^3$  particles and  $512^3$  grid cells. I use the flat model with  $\Omega_m = 0.4$  and  $\Omega_\Lambda = 0.6$ . To select the late-type galaxies, I simply pick dark matter particles at random. To select the early-type galaxies, I smooth the density field with a  $3 h^{-1}$  Mpc Gaussian filter, and apply a threshold of  $\delta_{c,e} = 0.25$ ; every dark matter particle above the threshold has an equal probability of becoming an early-type galaxy.

To create realistic mock catalogs, I pick a random particle in the simulation to represent the observer. I then “observe” the galaxies in the simulation box, using the angular and photometric limits of the LCRS, as well as the number of fibers available in each field (Shectman *et al.* 1996). Furthermore, there is a

Table 2. Properties of PDF and bias fits to realistic mock catalogs and volume-limited mock catalogs.

Catalog Type	$\sigma_e$	$\sigma_l$	Bias Model	$b_1$	$b_2$ or $\sigma_b$	$\mathcal{L}$
Volume-limited	$0.85 \pm 0.02$	$0.54 \pm 0.01$	Linear	$0.72 \pm 0.01$	—	0.0
			Quadratic	$0.74 \pm 0.01$	$-0.01 \pm 0.00$	-5.2
			Broken	$0.67 \pm 0.01$	$0.79 \pm 0.02$	-14.3
			Stochastic	$0.72 \pm 0.01$	$0.06 \pm 0.01$	-8.8
Flux-limited	$0.66 \pm 0.03$	$0.46 \pm 0.02$	Linear	$0.76 \pm 0.02$	—	0.0
			Quadratic	$0.77 \pm 0.02$	$-0.05 \pm 0.01$	-2.8
			Broken	$0.70 \pm 0.04$	$0.82 \pm 0.05$	-2.0
			Stochastic	$0.74 \pm 0.03$	$0.07 \pm 0.02$	-2.7

probability of failing to observe the galaxy which is a function of its magnitude, determined from the data. In the real LCRS, fibers could not be placed more closely than  $55''$ ; I implement that restriction in the mock catalogs as well. Finally, I include the appropriate magnitude and redshift errors.

As a benchmark, I take the simulation and divide it into cubic cells about  $25 h^{-1}$  Mpc on a side. I subsample the galaxies such that there are about 20–30 galaxies of each type in each cell. I refer to this sample as the volume-limited catalog. It is free of all of the selection effects associated with the real survey, as well as redshift-space distortions. The cells are equivalent in volume to the cells described in Section 3. I will evaluate the degree to which the selection effects affect the results by comparing realistic mock catalogs to the results for these volume-limited cells.

The results of this comparison are listed in Table 2. There is a significant difference between the variances measured for the volume-limited catalog and for the realistic mock catalog. I believe this is simply due to the fact that the non-cubical shapes of the cells in the LCRS allow the cells to probe effectively larger scales, thus driving down the variances. Tests show that the effect is not due to the incomplete sampling or due to fiber collisions. The important result, however, is that the bias is qualitatively the same (though with some quantitative differences). In particular, there is no increased  $\sigma_b$  for the mock catalogs; the ratio  $\sigma_b/\sigma_l$  is about 50% higher for the mock catalogs than the volume-limited catalogs, but this level is still low enough for us to conclude that the stochasticity I observe in the real catalog is not caused by any of the observational effects modeled here.

## 6. Conclusions

I have presented a powerful method of analyzing the nature of the relationship between different galaxy types. As a probe of stochasticity it is superior to calculating second moments; in addition, it can measure nonlinearity. It would be both simpler to implement and less susceptible to the problems of selection I have run into here if it were used to analyze a volume-limited survey, which I intend to do using the Optical Redshift Survey.

I have interpreted the redshift-dependence of the results as a problem in the understanding of the selection effects of the survey. The outstanding selection effect which I have *not* modeled yet in my mock catalogs is the central magnitude cut. I am in the process of determining whether the effect on the selection function is strong enough to cause the redshift-dependence found in Table 1. Note, importantly, that the results of Tegmark & Bromley (1999) are susceptible to the same problems I have found here, and should be interpreted accordingly.

In conclusion, the result of this work I take most seriously is the bottom quarter of Table 1, which excludes the two innermost rings, and indicates a bias which is linear, with perhaps some mild scatter, and an amplitude of  $b_1 \approx 0.8$ .

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